# **Buying Bitcoin - When The Low Comes Before The High**

#### **Overview:**

- Inspiration
- Data
- Model
- Results
- Future Work



# **Actual Inspiration and Intuition**

In a Past Project (similar stock predictions)

- Using:
  - Price History
  - Quarterly Financial Statements
  - Insider Trading History

Good for timing/trend indicator of a company's stock price

**The Question:** Can the same be done for cryptocurrency

## Can predictions be made without fundamentals?

#### Kind of...

- Things to keep in mind:
  - Stocks have metrics which investors take very seriously
  - Cryptocurrencies have no fundamental value
    - Some cryptos are tied to real economic value (compute services)
    - No standard method of comparing this real value provided and its relative worth to a "stable coin" like Bitcoin

## An Aside: Universal Price Formation Patterns in Order Books\*\*

### Credit where it is due. A lot was learned from this paper

- Data
  - Billions of sequential orderbook
- Goal
  - Prediction of next tick direction
- Findings
  - Universal patterns between industry and time periods on out of sample data
  - Analogous to language modeling

#### \*\* Citation:

Justin Sirignano & Rama Cont (2019) Universal features of price formation in financial markets: perspectives from deep learning, Quantitative Finance, 19:9, 1449-1459, DOI: 10.1080/14697688.2019.1622295



# BitMEX Exchange – What's Available

#### **Field Description** timestamp - time (microsecond) symbol contract (exchange handles more than BTC) Side - buy/sell - # of contracts = USD size - Price of Contract Price - ['MinusTick', 'ZeroMinusTick', 'ZeroPlusTick', 'PlusTick'] tickDirection grossValue - Number of satoshis 1 satoshi = 0.00000001 units Home Notional - # of coin

- # of US dollars

Foreign Notional

## **How Data Is Used**

### **Field**

## **Description**

- timestamp
- symbol
- Side
- size
- Price
- Tick Direction

- Collect trades into 1 minute intervals for analysis
- include trade in analysis
- buy/sell. Important filter for other metrics
- Summary statistics based on volume
- Price of Contract
- Counts per time period (inspired by price formation paper)

# **Further Processing**

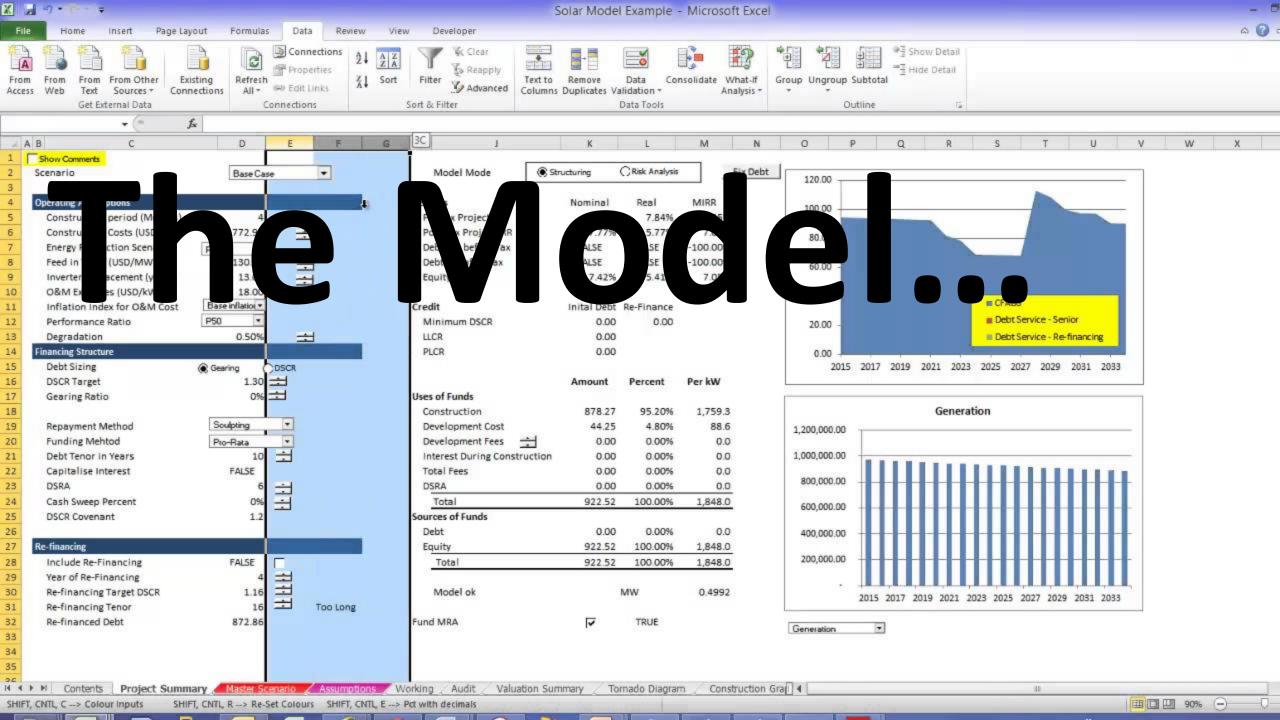
Returns Relative to Prior VWAP

$$Minute\_Min\_Vs\_VWAP = \frac{PriceMin_t - PriceMin_{t-1}}{VWAP_{t-1}}$$

Returns Relative to Own Metric

$$Minute\_Min\_Vs\_Self = \frac{PriceMin_t - PriceMin_{t-1}}{PriceMin_{t-1}}$$

About 40 variations experimented with... kept 10 most valuable



# At Midterm Presentation: Model was too big / looking too far forward

Out[93]:

In [93]: df.head()

Data Imports as: ~ 500k trades/day

	symbol	side	size	price	tickDirection	trdMatchID	grossValue	homeNotional	foreignNotiona
timestamp									
2017-08-18 00:00:01.536113	XBTUSD	Buy	202	4278.3	ZeroPlusTick	fd8cc9de-9e76-a2ce-196c-1e5e64bedeee	4721548	0.047215	202.0
2017-08-18 00:00:01.536113	XBTUSD	Buy	909	4278.4	PlusTick	3c4f3bb7-c431-22b9-033b-4e769b506dce	21246057	0.212461	909.0
2017-08-18 00:00:02.160534	XBTUSD	Sell	398	4278.2	MinusTick	e05c6ef3-ae5b-e001-e253-324ece1b9133	9302852	0.093029	398.0
2017-08-18 00:00:02.160534	XBTUSD	Sell	11643	4278.1	MinusTick	a815b31a-9cfa-b6ef-b2fe-b5c074887561	272155125	2.721551	11643.0
2017-08-18 00:00:02.472558	XBTUSD	Buy	1988	4278.3	PlusTick	f9e800b2-72ff-5e47-9c3f-f789e9ce91ae	46467512	0.464675	1988.0

#### **Transformed to:**

Observation #	Χ	(feature:	s)	Y (target)			
1	$x_1$	•••	$x_t$	$y_{t+1}$	•••	$y_{t+z}$	
	•••	•••	•••		•••	•••	
n	$x_n$	•••	$x_{n+t-1}$	$y_{t+n}$	•••	$y_{t+n+z}$	
			,	•			

47 minute lookback

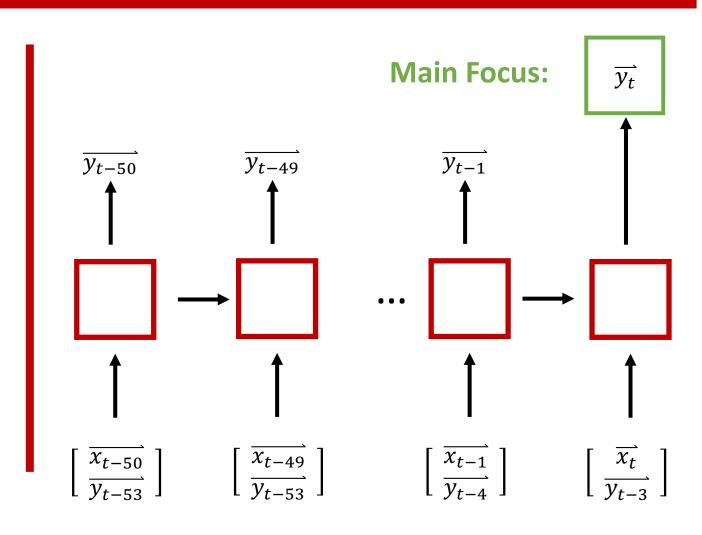
3 minute future prediction

## Visualization of Final Model's Flow

Features:  $\overrightarrow{x_t} \in \mathbb{R}^{21}$ 

**Targets:**  $\overrightarrow{y_t} \in \mathbb{R}^6$ 

**Model Inputs:**  $\left[\begin{array}{c} \overrightarrow{x_t} \\ \overline{y_{t-3}} \end{array}\right] \in \mathbb{R}^{27}$ 



## **Back to Basics: Summary of How Observations Are Made**

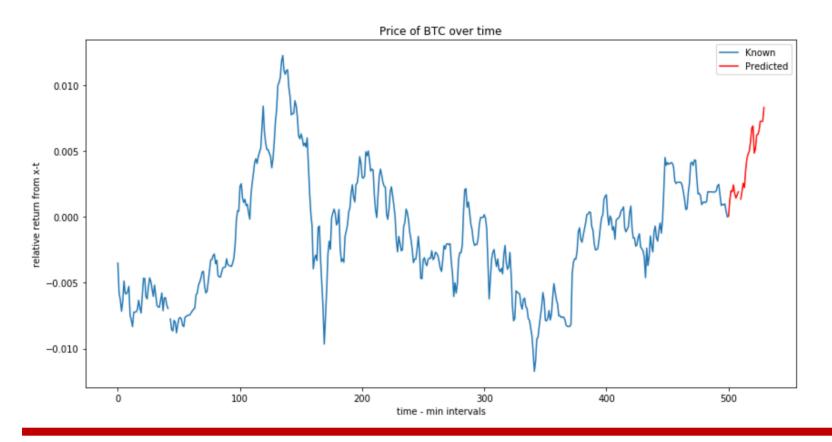
	2	<b>X</b> (features	5)	Y (target)					
Observation:	$x_1$	•••	$x_t$	$y_{t+1}$	•••	$y_{t+z}$			
<b>†</b>									
Currently Prices									
<b>↓</b>									
Transform To returns by									

		)	<b>K</b> (features	)	<b>Y</b> (target)		
Observation	on:	$\frac{x_1 - x_t}{x_t}$		$\frac{x_t - x_t}{x_t}$	$\frac{y_{t+1} - x_t}{x_t}$		$\frac{y_{t+z} - x_t}{x_t}$

# (Last Time) Price Prediction

## **Primary Goal:**

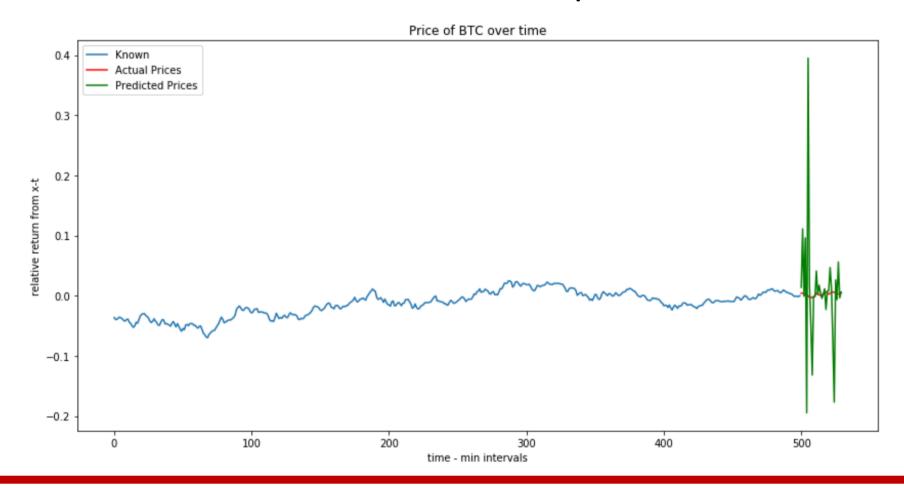
Predict Future Returns In Short Term



Unfortunately...
LSTMs can be tricky

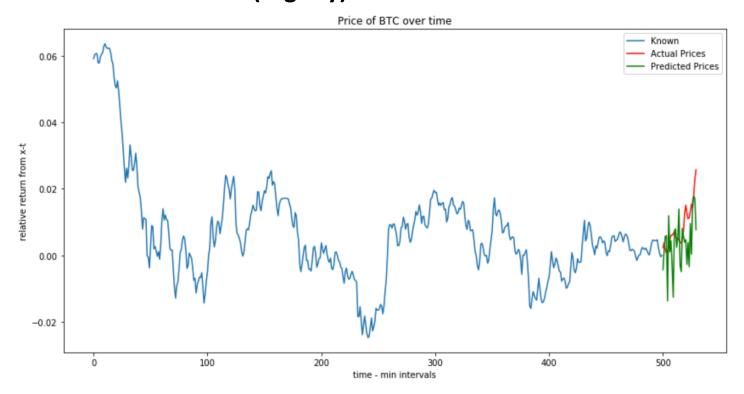
# (Last Time) Predictions Have Looked More Like This:

#### LSTM Predictions... much more volatile than actual prices



# (Last Time) With Some Hyper-Parameter Tuning

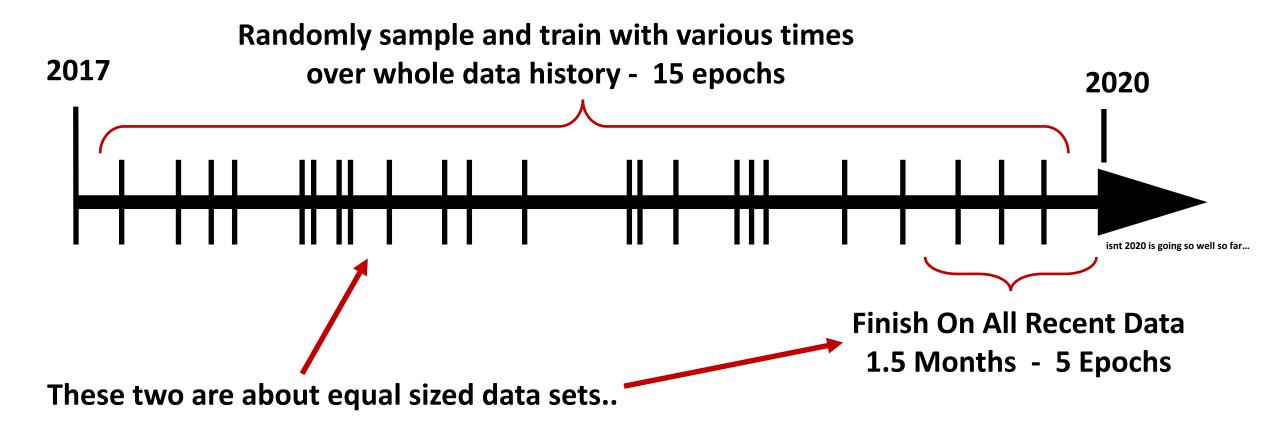
#### We Can Fix This Problem (slightly)



Prediction volatility persists enough...

Maybe good to explore alternatives to price timeseries forecasting

## So What Is Different Now?

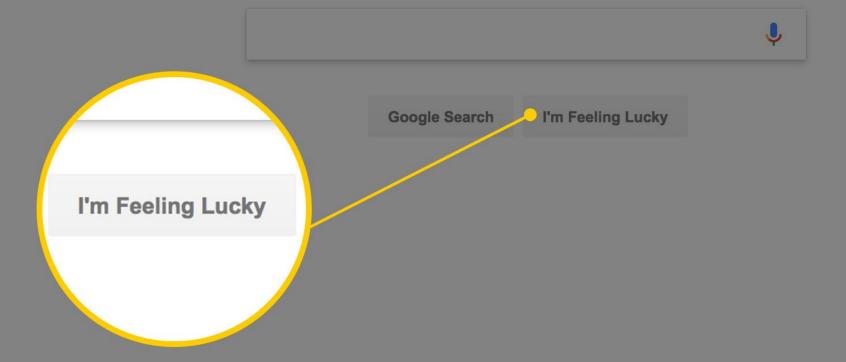


<sup>\*\*</sup> Inspired by the Universal Price Formation Paper

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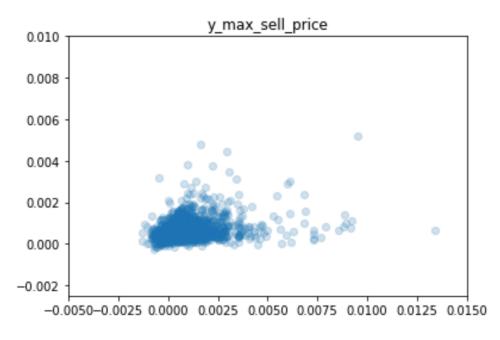
# Results

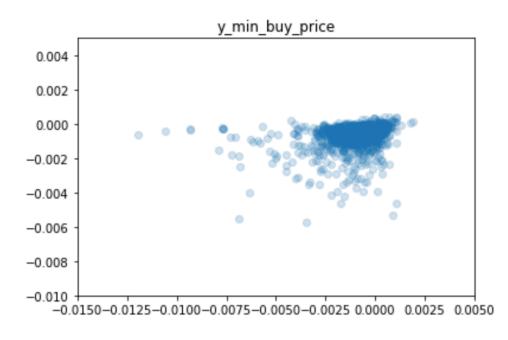
# Google



## **Final Version Predictions: Relative Prices**

- Given Model Input Observe Final  $y_t$ 
  - Below: min and max price relative to prior interval's closing price



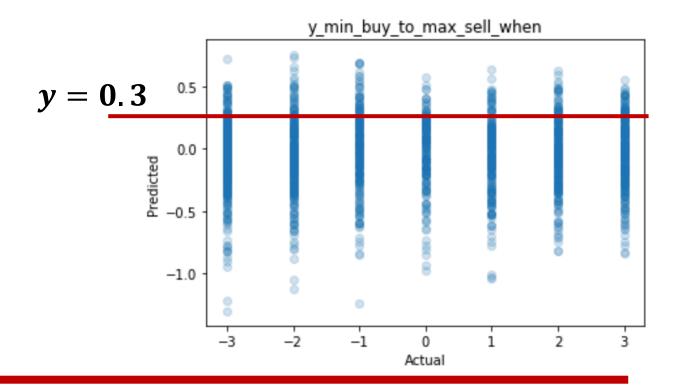


These results look like they could be promising, but they're missing one key factor

# Final Version Predictions: Timing

- Initially timing does not seem as useful
- ... combining with some hard-coded rules makes results are more promising

Consider only trades where predicted min to max timing is Greater than 0.3...

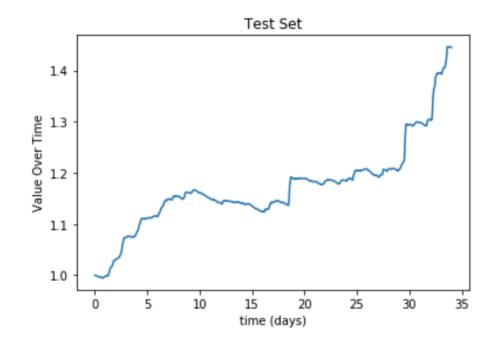


# Hard Coding is Important – a typical run

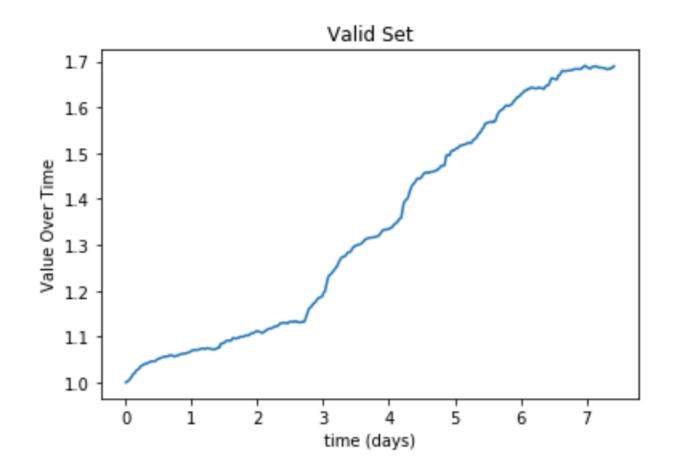
#### **Optimal Procedure from some tuning**

- If min to max timing > 0.3
- Place limit order for 0.993 of min prediction price
- Set Immediate Stop Loss of 0.15% on trade (assume if necessary it will hit within 2 ticks on average, a tick is equal to a penny)
- If reaches max predicted price set stop loss at that price, let run till end of 3 minute period.
  - As price rises adjust stop loss up so it hits 3 pennies below current price

## **Results**



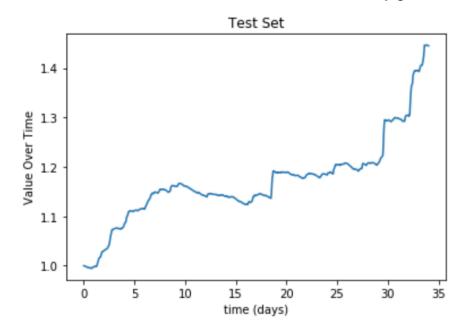
## **Best Run**



## If Its too good to be true... On Average Run

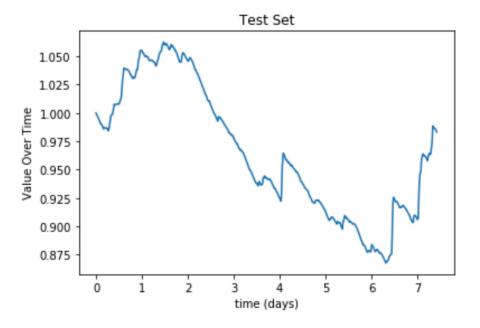
#### **Actual Trading Fees**

•  $0.000750249 \approx 0.075 \%$ 



#### **Raised Slightly Trading Fees**

•  $0.0012496 \approx 0.12496 \%$ 

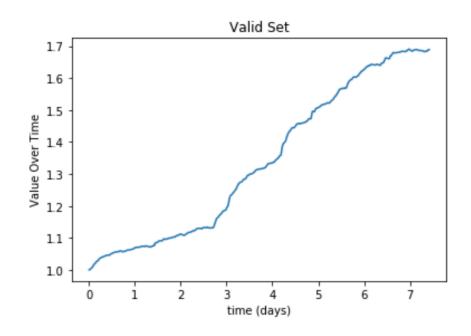


Results will be somewhere in between due to exchange's fee structure

## If Its too good to be true... On Best Run

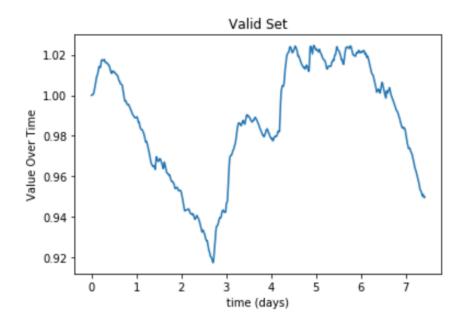
#### **Actual Trading Fees**

•  $0.000750249 \approx 0.075 \%$ 



#### **Raised Slightly Trading Fees**

•  $0.0012496 \approx 0.12496 \%$ 



The right side is a worst case scenario under some (hopefully) light assumptions

## Summary

## Approach / What Worked

- Data Cleaning Process
- Price predictions over time are hard
  - But...
- Current model structure
  - Relative min / max and when a bit easier

## **Uncertainty**

- Are assumptions reasonable
  - Trading fees especially
- Looking at BitMEX volume it seems that hitting within 2 ticks of a set limit is reasonable
- Communication time due to inexperience your guess is as good as mine